

# efficiency scale factor for bs and mistag estimate

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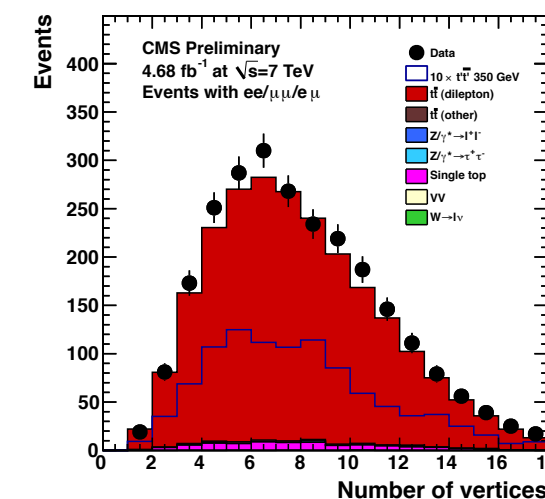
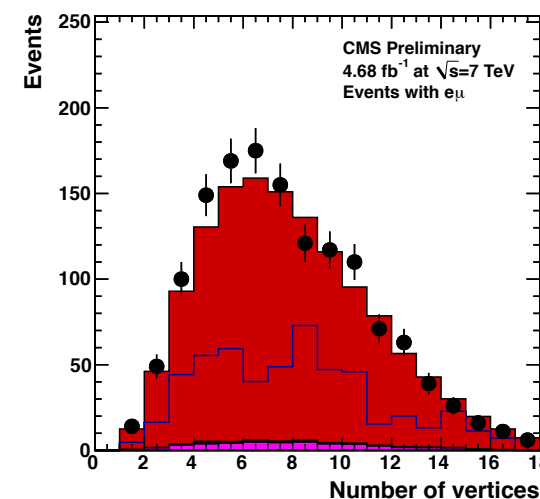
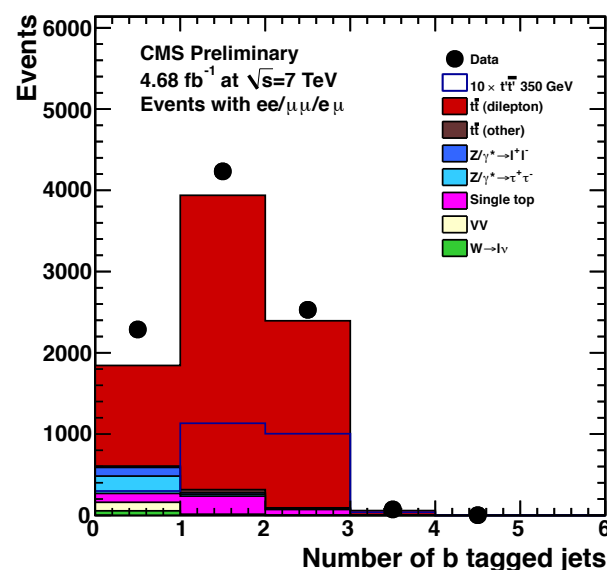
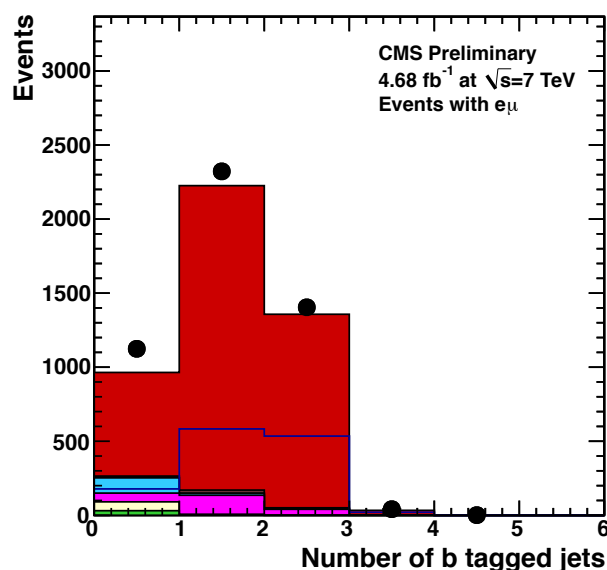
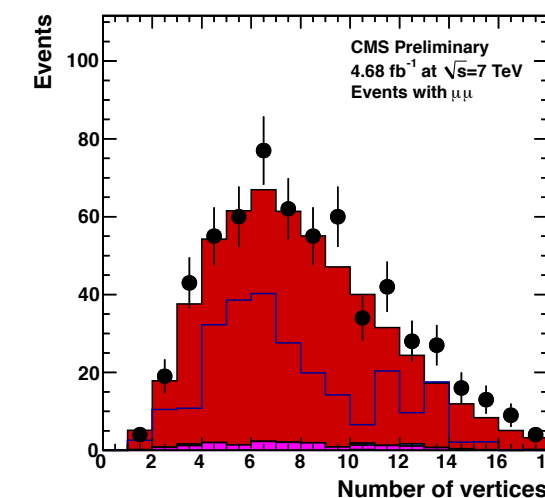
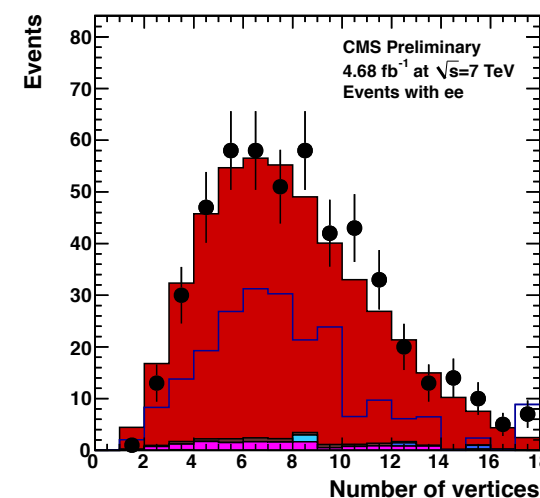
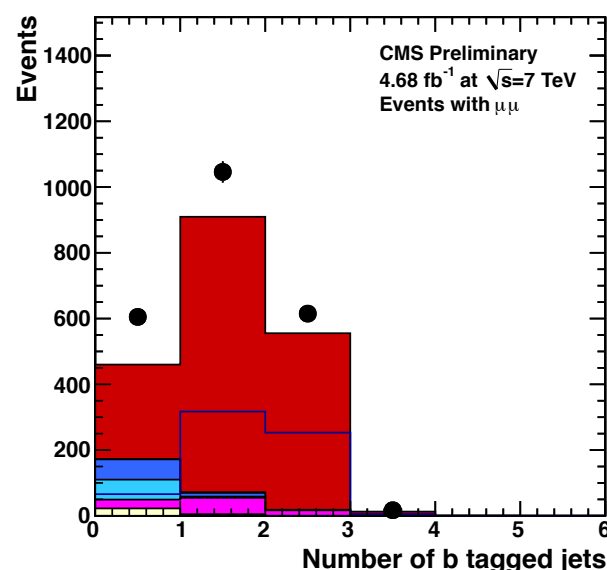
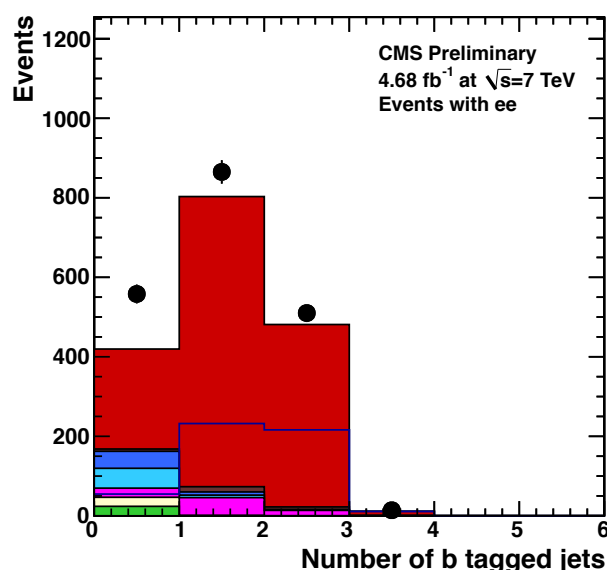
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$t'$  meeting

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I showed a plot a month ago with bad data-MC agreement in the 0 and 1 b-tag jet bins. The 2 b-tag bin is our preselection region.



Plot uses 4.7/fb and the new large Fall I ttbar sample

new vertex weighting: #vertex distribution looks OK (right plots)

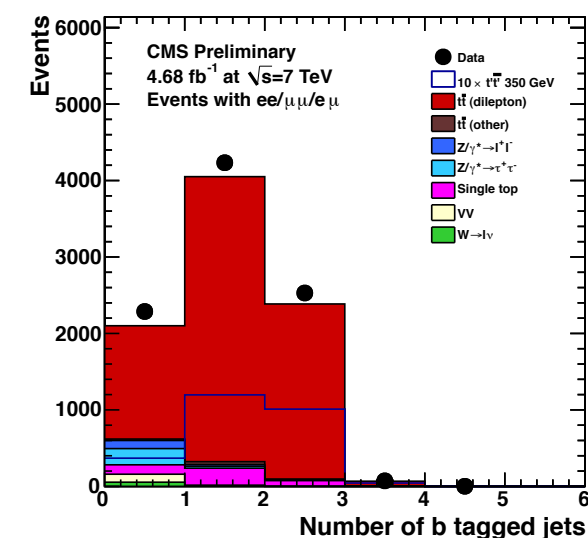
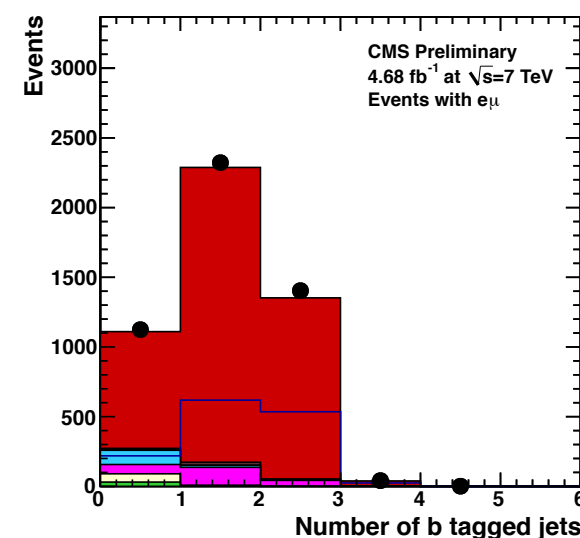
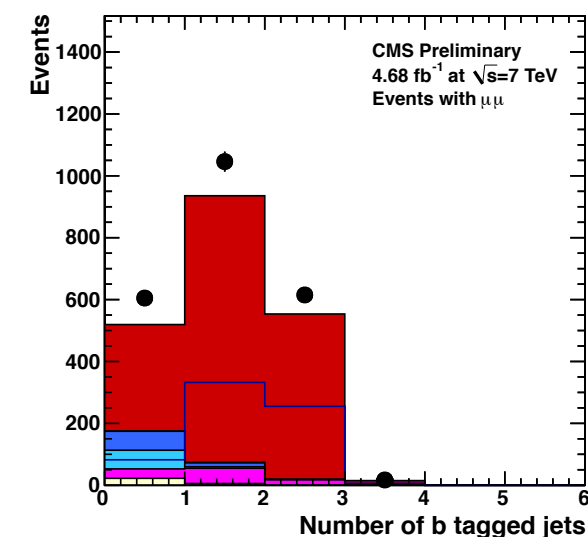
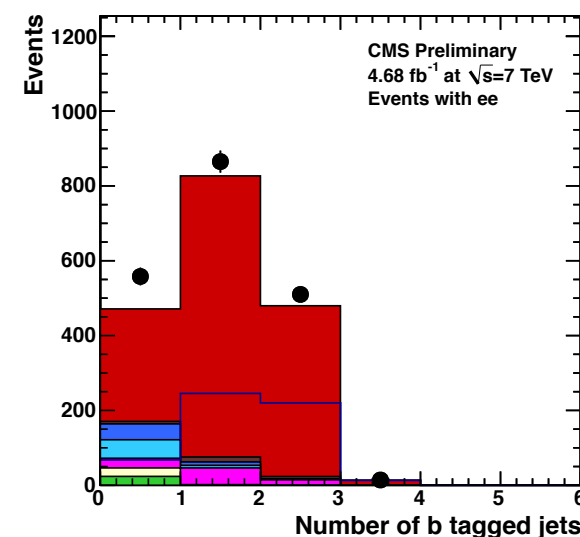
- ▶ Realised that our btag efficiency weighting is too much of an approximation
  - ▶ we use a weight of 0.95 for each btagged jet, where 0.95 is the data/MC scale factor for b-tagging efficiency. This gives a weight of  $0.95^2$  for events in the signal region.
  - ▶ In fact, we should only weight the **real bs** that were tagged.
  - ▶ For the remaining b-tagged jets (mistags) the factor of 0.95 is not correct, and we should use the data/MC scale factor for mistags instead ( $\sim 1.25$ ).
  - ▶ Furthermore, we mustn't neglect the real bs that were **not** tagged
    - ▶ A data-MC scale factor is also applicable here. For example, an event with 2 real bs but 0 b tags is less likely to appear in MC than in data due to the higher tagging efficiency in MC
  - ▶ Finally, there is a similar scale factor for the mistagable jets that were not mistagged.
- ▶ Thus, overall the weight for each event comes from the product of the four scale factors: real b to be tagged \* real b not tagged \* light jet mistagged \* light jet not mistagged

Num. b-tagged jets matched to real b    Num. untagged jets matched to real b    Num. b-tagged jets not matched to real b    Num. untagged jets not matched to real b

$$F_t^{N_t^b} \times F_{nt}^{N_{nt}^b} \times F_m^{N_m^j} \times F_{nm}^{N_{nm}^j}$$

- ▶ These scale factors are taken from <http://cms-physics.web.cern.ch/cms-physics/public/BTV-11-001-pas.pdf>
- ▶ The MC #btags distribution matches the data better when using this weighting

- ▶ #btag distributions after weighting
- ▶ Note, overall yield with Fall11 madgraph ttbar is 3.3% lower than with Summer11 ttbar
- ▶ not sure why
- ▶ Table compares total MC yield to data with the old and new btag efficiency weighting, using either the Fall11 or Summer11 ttbar madgraph sample
- ▶ Better agreement with new weighting
- ▶ Better agreement with Summer11 ttbar



#btags	Fall11, 0.95 <sup>n</sup>	Fall11, new	Sum11, 0.95 <sup>n</sup>	Sum11, new	data
0	1,844.0	2,100.8	1,911.1	2,179.1	2287
1	3,741.4	4,050.6	3,889.6	4,208.7	4233
2	2,393.8	2,385.6	2,453.1	2,445.2	2529
3	54.9	67.3	55.1	67.4	72
4+	2.0	2.8	1.1	1.7	1
All	8,036.1	8,607.1	8,310.0	8,902.2	9122



► Signal region yields using Fall I I madgraph ttbar sample

Sample	ee	$\mu\mu$	$e\mu$	all
ttprime350	$4.9091 \pm 0.9196$	$7.2041 \pm 1.0244$	$10.8035 \pm 1.2854$	$22.9166 \pm 1.8834$
ttprime400	$3.1642 \pm 0.4660$	$4.8099 \pm 0.5610$	$9.9882 \pm 0.8301$	$17.9624 \pm 1.1050$
ttprime450	$2.1583 \pm 0.2781$	$2.3028 \pm 0.2698$	$5.8573 \pm 0.4382$	$10.3185 \pm 0.5850$
ttprime500	$1.3145 \pm 0.1526$	$1.7012 \pm 0.1645$	$2.9942 \pm 0.2202$	$6.0099 \pm 0.3143$
ttprime550	$0.9210 \pm 0.1029$	$0.9156 \pm 0.0944$	$1.8336 \pm 0.1373$	$3.6702 \pm 0.1959$
ttprime600	$0.5142 \pm 0.0568$	$0.5265 \pm 0.0544$	$1.1911 \pm 0.0833$	$2.2318 \pm 0.1146$
ttdil	$0.1659 \pm 0.0547$	$0.1121 \pm 0.0446$	$0.2719 \pm 0.0704$	$0.5499 \pm 0.0997$
ttotr	$0.0137 \pm 0.0123$	$0.0113 \pm 0.0113$	$0.0355 \pm 0.0223$	$0.0605 \pm 0.0278$
wjets	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$
DYee	$0.6203 \pm 0.6203$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.6203 \pm 0.6203$
DYmm	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$
DYtautau	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$
VV	$0.0000 \pm 0.0000$	$0.0085 \pm 0.0066$	$0.4092 \pm 0.2894$	$0.4177 \pm 0.2895$
tw	$0.0000 \pm 0.0000$	$0.1194 \pm 0.0849$	$0.3620 \pm 0.1629$	$0.4814 \pm 0.1837$
Total MC	$0.7999 \pm 0.6228$	$0.2513 \pm 0.0968$	$1.0787 \pm 0.3402$	$2.1299 \pm 0.7163$
Data	$0.0000 \pm 0.0000$	$0.0000 \pm 0.0000$	$1.0000 \pm 1.0000$	$1.0000 \pm 1.0000$

► Supports earlier results: ttbar no longer dominant background

Prediction of the method compared to true count of events with mistags in MC for each sample

MC Sample	mistag bkg prediction	true # mistag events	pred/true
tt dil (Fall11 madgraph)	$0.28 \pm 0.02$	$0.45 \pm 0.09$	61%
tt other (Fall11 madgraph)	$0.03 \pm 0.01$	$0.05 \pm 0.03$	51%
w+jets	$0.00 \pm 0.01$	0	
DYee	$0.27 \pm 0.13$	0	
DYmm	$0.04 \pm 0.04$	0	
DYtautau	$0.30 \pm 0.16$	0	
VV	$0.02 \pm 0.01$	$0.41 \pm 0.29$	5%
tw	$0.12 \pm 0.02$	$0.48 \pm 0.18$	25%
<b>Sum of MC</b>	<b><math>1.04 \pm 0.21</math></b>	<b><math>1.40 \pm 0.36</math></b>	<b>74%</b>
tt dil (powheg)	$0.31 \pm 0.02$	$0.51 \pm 0.08$	61%

Total prediction covers 74% total mistags, but much less for VV and tW

Total MC yield (all events):  $2.13 \pm 0.72$  (i.e.  $1.40/2.13=66\%$  are from mistags)

82% total Fall11 tt yield is from mistags

prediction covers  $1.04/2.13 = 49\%$  of total background

Mistag method background prediction in data:  **$0.74 \pm 0.27$**

► Same as last slide, but for preselection region (without  $M_{lb}^{\min}$  cut)

Sample	mistag bkg prediction	true # mistag events	pred/true	
ttdil	131.62 $\pm$ 0.33	233.90 $\pm$ 1.95	0.56 $\pm$ 0.00	
ttotr	1.62 $\pm$ 0.04	2.69 $\pm$ 0.21	0.60 $\pm$ 0.05	
wjets	0.094 $\pm$ 0.101	0.00 $\pm$ 0.00		
DYee	0.66 $\pm$ 0.23	0.00 $\pm$ 0.00		
DYmm	0.65 $\pm$ 0.30	0.00 $\pm$ 0.00		
DYtautau	0.94 $\pm$ 0.26	1.67 $\pm$ 0.97	0.56 $\pm$ 0.36	
VV	0.24 $\pm$ 0.04	1.01 $\pm$ 0.41	0.24 $\pm$ 0.11	
tw	7.05 $\pm$ 0.15	25.39 $\pm$ 1.27	0.28 $\pm$ 0.02	
sum of MC	142.87 $\pm$ 0.60	264.66 $\pm$ 2.56	0.54 $\pm$ 0.01	
Data	137.70 $\pm$ 2.71	N/A	0.05 $\pm$ 0.00	

► Prediction covers 54% total mistags

► pred/true consistent with previous slide for the individual samples

► Data prediction matches the sum of MC predictions

► Overall pred/true was 74% in signal region: is this a biased result due to 0 mistag DY events in signal region in MC?

- ▶ Method works reasonably well, but final prediction covers 50% (or less) of total background
- ▶ We can live with this because the effect of a 50% underestimation of background on our limit is only  $\sim 10$  GeV
- ▶ Assign 100% syst. uncertainty

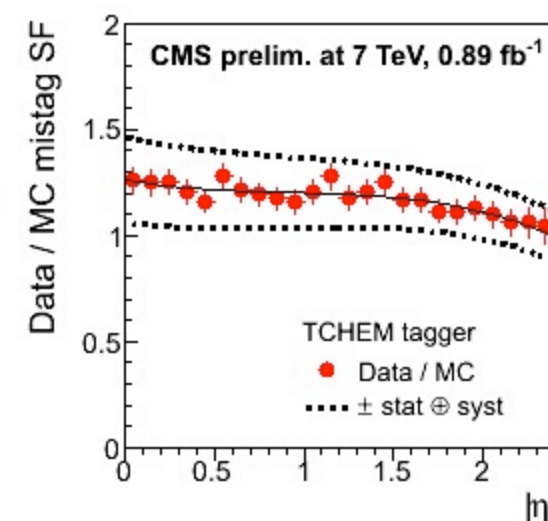
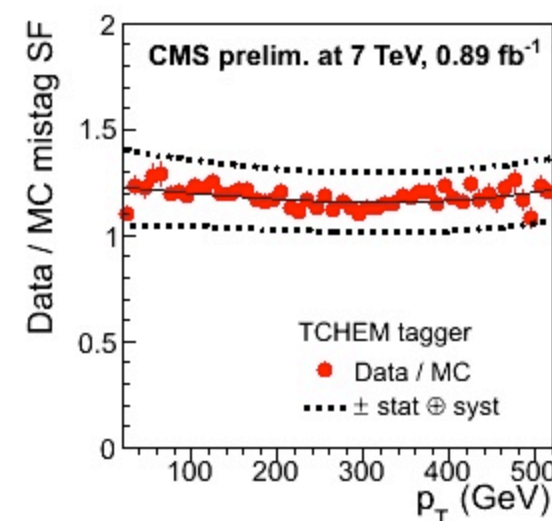
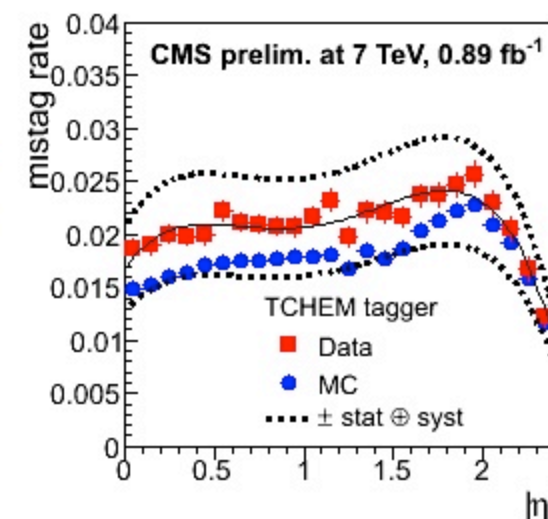
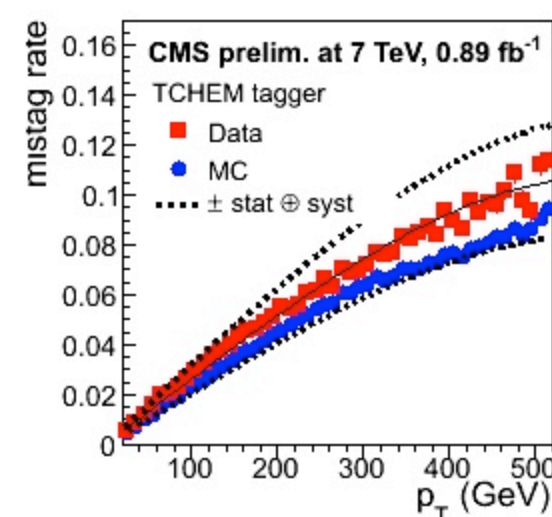




- ▶ I'm not yet entirely happy with the efficiency weighting
- ▶ real b tags are defined as b-tagged jets that match a real b within  $dR < 0.4$  (and each real b is only allowed to match 1 btagged jet)
- ▶ all other b tagged jets are assumed to be mistags, and weighted as appropriate
  - ▶ so they pick up a weight of  $\sim 1.25$  instead of  $\sim 0.95$  (from SF plot below)
- ▶ Is it OK to call everything else a mistag like this? Some will be badly mismeasured bs.
- ▶ To test the effect of the mistag part of the weighting, I set the mistag scale factors to 1 (i.e. only used the 2 real b scale factors).

Results are similar:  
mistag SFs overall  
effect small except  
with  $\geq 3$  btags

#btags	b SFs	all SFs
0	2111.1	2100.8
1	4022.6	4050.6
2	2345.0	2385.6
3	57	67.3
4+	2.1	2.8
All	8537.3	8607.1



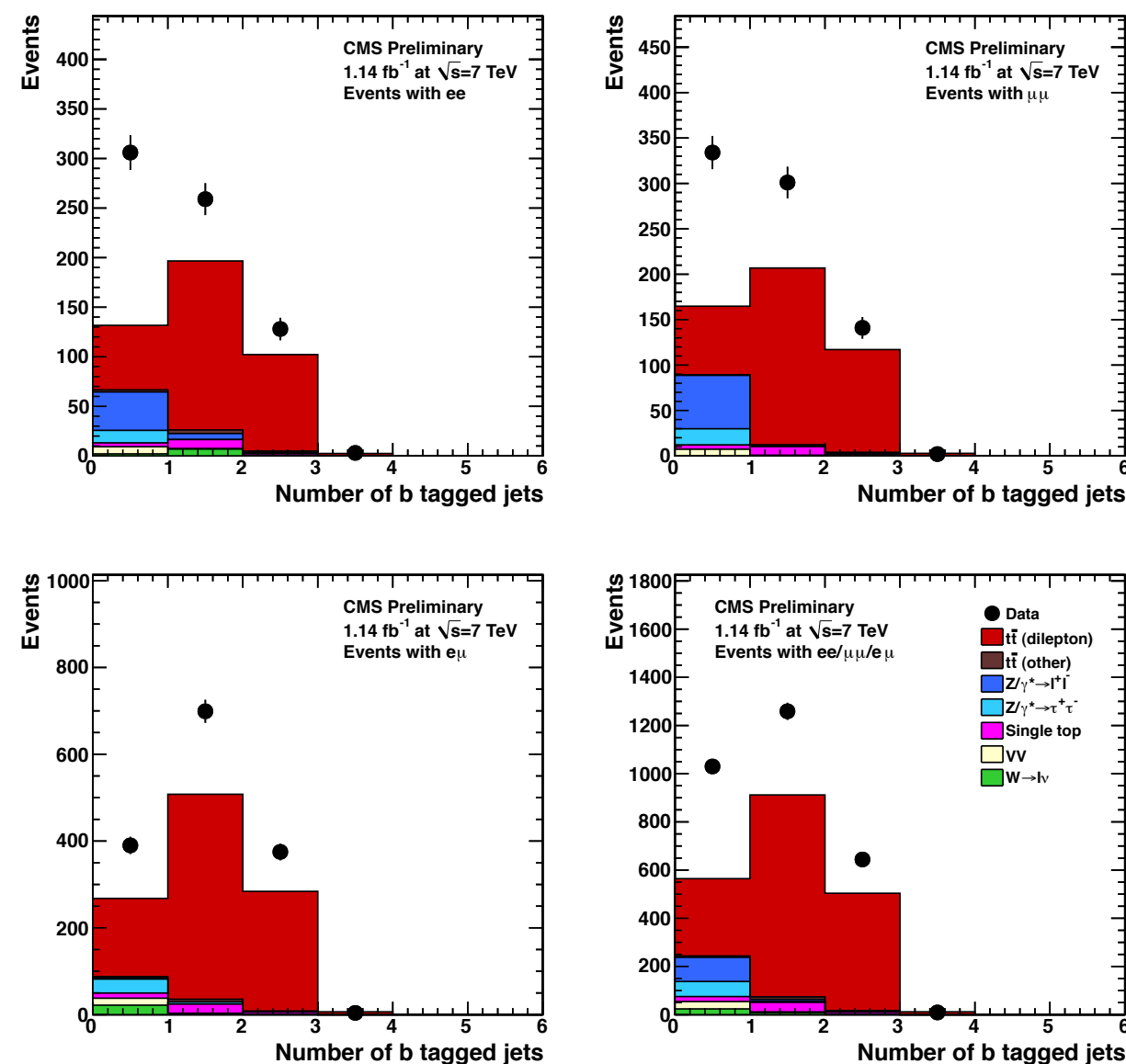
- ▶ The 1-btag bin data yields now match the MC much better
- ▶ Consider using the 1-btag yields to estimate 2 b-tag yields?
  - ▶  $N_{\text{predicted}}^{2\text{btags}} = N_{\text{data}}^{1\text{btag}} * R_{\text{mc}}$ 
    - ▶  $R_{\text{mc}} = N_{\text{mc}}^{2\text{btags}} / N_{\text{mc}}^{1\text{btag}}$
  - ▶  $N_{\text{predicted}}^{2\text{btags}} = 17 * (2.13/16.69) = 2.17 \pm 0.76$
- ▶ But  $R_{\text{mc}}$  might not be well modelled
- ▶ And very poor MC stats ( $N_{\text{mc}}^{2\text{btags}} = 2.13 \pm 0.72$ )

- Prediction of the method compared to true count of events with mistags in MC for each sample (3.23/fb, Summer11 MC)

MC Sample	mistag bkg prediction	true # mistag events
tt dil (madgraph)	0.22 $\pm$ 0.06	0.16 $\pm$ 0.16
tt other	0.002 $\pm$ 0.002	0
w+jets	0.000 $\pm$ 0.004	0
DYee	0.13 $\pm$ 0.06	0
DYmm	0.02 $\pm$ 0.02	0
DYtautau	0.13 $\pm$ 0.06	0
VV	0.01 $\pm$ 0.01	0.17 $\pm$ 0.12
tw	0.06 $\pm$ 0.01	0.25 $\pm$ 0.10
<b>Sum of MC</b>	<b>0.57 <math>\pm</math>0.11</b>	<b>0.58 <math>\pm</math>0.22</b>
also looked at powheg pythia tt sample (better stats):		
tt dil (powheg)	0.16 $\pm$ 0.01	0.25 $\pm$ 0.04

- reasonable closure for ttbar and also for sum of MC, but statistics are very limited

► This is the actual plot I showed 1 month ago



- Two reasons it's even worse than the plot on slide 1:
- weight of 0.95<sup>2</sup> for all events (slide 1 has 0.95<sup>2</sup> for  $\geq 2$ , and 1.00 otherwise)
  - powheg ttbar (overall lower yield than madgraph)